

# General Clique Percolation in Network Evolution

Jingfang Fan<sup>1</sup> and Xiaosong Chen<sup>1,\*</sup>

<sup>1</sup> State Key Laboratory of Theoretical Physics, Institute of Theoretical Physics, Chinese Academy of Sciences, P.O. Box 2735, Beijing 100190, China

We introduce a general  $(k, l)$  clique community, which consists of adjacent  $k$ -cliques sharing at least  $l$  vertices with  $k - 1 \geq l \geq 1$ . The emergence of a giant  $(k, l)$  clique community indicates a  $(k, l)$  clique percolation, which is studied by the largest size gap  $\Delta$  of the largest clique community during network evolution and the corresponding evolution step  $T_c$ . For a clique percolation, the averages of  $\Delta$  and  $T_c$  and the root-mean-squares of their fluctuations have power law finite-size effects whose exponents are related to the critical exponents. The fluctuation distribution functions of  $\Delta$  and  $T_c$  follow a finite-size scaling form. In the evolution of the Erdős-Rényi network, there are a series of  $(k, l)$  clique percolation with  $(k, l) = (2, 1), (3, 1), (3, 2), (4, 1), (4, 2), (5, 1), (4, 3)$ , and so on. The critical exponents of clique percolation depend on  $l$ , but are independent of  $k$ . The universality class of a  $(k, l)$  clique percolation is characterized alone by  $l$ .

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Community structure is of great interest in the studies of networks [1, 2]. The term network community is defined as a group of vertices that are more densely connected each other than other vertices in a network. The clique [3] and core [4] are the two examples of network community. To analyze the overlapping community structure of networks, Palla *et al.*[5] proposed the clique percolation method (CMP) to build up the communities from  $k$ -cliques, which is a fully connected subset of  $k$  vertices. Two  $k$ -cliques are considered to be adjacent if they share  $k - 1$  vertices. A clique community is defined as the maximal union of  $k$ -cliques that can be reached from each through a series of adjacent  $k$ -cliques.

When the size of a clique community is comparable to network size  $N$ , a giant clique community emerges and there is a clique percolation. Derényi *et al.* [6] studied the clique percolation of the Erdős-Rényi (ER) model[7]. The  $k$ -cliques percolation takes place when the probability of connecting two vertices in the network reaches the threshold  $p_c(k) = [(k - 1)N]^{-1/(k-1)}$  [6]. The normal percolation transition of ER model corresponds to the  $k = 2$  clique percolation.

In this Letter, we introduce a general  $(k, l)$  clique community, where any two adjacent  $k$ -cliques share at least  $l$  vertices with  $k - 1 \geq l \geq 1$ . For the same network, the clique communities of different  $(k, l)$  are different. As an illustration,  $(3, 1)$  and  $(3, 2)$  clique communities in a network are shown in Fig.1 and they are different. A  $(k, l)$  clique percolation appears with the emergence of a giant  $(k, l)$  clique community. The previous clique percolation discussed in Ref.[6] corresponds to the special case  $l = k - 1$  of general  $(k, l)$  clique percolation. The normal percolation transition corresponds to a  $(2, 1)$  clique percolation. Here we study the general  $(k, l)$  clique percolation of the ER model by analyzing the finite-size effects of network evolution. From the power law exponents of finite-size effects, we can obtain the critical exponents of  $(k, l)$  clique percolation.

At first, we calculate the threshold of the  $(k, l)$  clique percolation using the approach developed by Newman, Watts,

and Strogatz[9]. From the probability distribution  $p_k$  of vertex degrees, we define the generating functions  $G_0(x) = \sum_{k=0}^{\infty} p_k x^k$  and  $G_1(x) = \sum_{k=0}^{\infty} q_k x^k$ , where  $q_k = p_{k+1}(k + 1)/<k>$  is the excess degree distribution of network [10].

The generating function  $H_1(x)$  of the probability distribution for the sizes of the components reached from a randomly chosen edge satisfies a self-consistent equation

$$H_1(x) = xG_1(H_1(x)). \quad (1)$$

When its first derivative  $H'_1(1)$  becomes infinite, a giant component appears and there is a percolation. This corresponds to  $z' = G'_1(1) = 1$ .

After taking  $k$ -clique as the new unit and considering two cliques to be adjacent if they share at least  $l$  vertices, the average degree of a  $k$ -cliques is

$$z = \sum_{k'=l}^{k-1} \binom{N-k}{k-k'} \binom{k}{k'} p^{\binom{k}{2} - \binom{k'}{2}}, \quad (2)$$

where  $p$  is the probability to connect two vertices with an edge. Correspondingly, the average excess degree of the  $k$ -

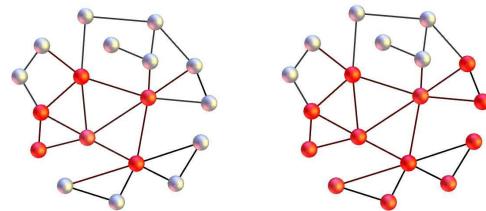


FIG. 1: (color online) Sketches of different clique communities in a network. The vertices belonging to a clique community are marked with red color. The largest  $(3, 2)$  clique community is shown on the left and the largest  $(3, 1)$  clique community on the right.

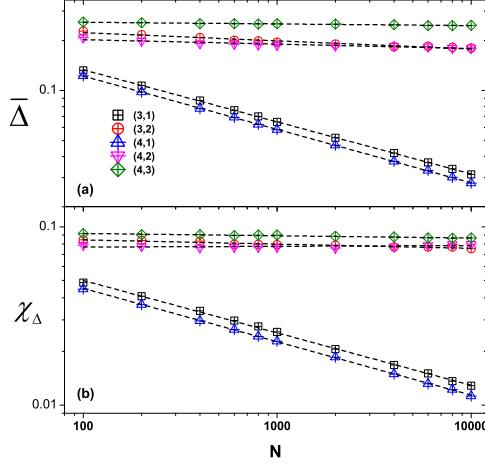


FIG. 2: (color online) Log-log plot of the average size gap  $\bar{\Delta}$  and the root-mean-square of its fluctuation  $\chi_{\Delta}$  versus  $N$  for different  $(k, l)$  clique percolation. Power law behaviors of  $\bar{\Delta}$  and  $\chi_{\Delta}$  are confirmed by simulation data. From the slopes of fitting lines,  $\beta_1$  and  $\beta_2$  can be obtained and are summarized in Table I.

cliques is

$$z' = \sum_{k'=l}^{k-1} \binom{N-k}{k-k'} \left[ \binom{k}{k'} - 1 \right] p^{\binom{k}{2}} - \binom{k'}{2}. \quad (3)$$

The threshold  $p_c(k, l)$  of  $(k, l)$  clique percolation can be determined from the equation

$$\sum_{k'=l}^{k-1} \frac{\binom{k}{k'} - 1}{(k-k')!} [1 + O(N^{-1})] \left[ N p_c^{\frac{k+k'-1}{2}} \right]^{k-k'} = 1, \quad (4)$$

which gives

$$p_c(k, l) = a(k, l) N^{-\frac{2}{k+l-1}} \left[ 1 + O(N^{-\frac{k-l-1}{k+l-1}}) \right], \quad (5)$$

$$a(k, l) = \left[ \frac{\binom{k}{l} - 1}{(k-l)!} \right]^{-\frac{2}{(k-l)(k+l-1)}}. \quad (6)$$

For  $l = k-1$ , the correction term in Eq.5 vanishes and  $p_c(k, k-1) = [(k-1)N]^{-\frac{1}{k-1}}$ , which is in agreement with the result of Derényi et al [6, 11]. For  $l < k-1$ , the correction term  $O(N^{-\frac{k-l-1}{k+l-1}})$  exists. For general  $(k, l)$ , only  $p_c(k, l) \sim N^{-\frac{2}{k+l-1}}$  is obtained by Bollobás and Riordan[8].

In the following, we study  $(k, l)$  clique percolation of ER model with Monte Carlo simulation and finite-size effects of network evolution [14]. During an evolution process, the largest  $(k, l)$  clique community has a size gap  $S_1(T) - S_1(T-1)$  at an evolution step  $T$ . The largest reduced size gap during the whole evolution process is

$$\Delta \equiv \frac{1}{N} \max \{ \{ S_1(T) - S_1(T-1) \} \}, \quad (7)$$

which could be related to a percolation transition at the corresponding evolution step  $T_c$ . From the results of  $\Delta$  and  $T_c$  in

many Monte Carlo simulations, we can calculate the average size gap  $\bar{\Delta}$  and the average transition point  $\bar{T}_c$ . Since  $T = p \times N(N-1)/2$  and therefore  $\bar{T}_c \propto N^{2-\frac{2}{k+l-1}}$ , it is convenient to introduce a reduced evolution step  $r = T/N^{2-\frac{2}{k+l-1}}$  and the reduced transition point  $r_c \equiv T_c/N^{2-\frac{2}{k+l-1}}$ . We anticipate that  $\bar{\Delta}$  and  $\bar{r}_c$  have the power law finite-size effects as

$$\bar{\Delta}(N) \sim N^{-\beta_1}, \quad (8)$$

$$\bar{r}_c(N) - r_c(\infty) \sim N^{-1/\nu_1}. \quad (9)$$

Using  $p_c(k, l)$  in Eq. 5, we get

$$r_c^T(\infty) = \frac{1}{2} a(k, l). \quad (10)$$

The character of clique percolation is determined by the exponent  $\beta_1$ . The clique percolation is continuous when  $0 < \beta_1 < 1$  and discontinuous when  $\beta_1 = 0$  [15].

The fluctuations  $\delta\Delta = \Delta - \bar{\Delta}(N)$  and  $\delta r_c = r_c - \bar{r}_c(N)$  are investigated also. Their root-mean-squares are defined as

$$\chi_{\Delta} = \sqrt{\langle [\delta\Delta]^2 \rangle}, \quad (11)$$

$$\chi_r = \sqrt{\langle [\delta r_c]^2 \rangle}, \quad (12)$$

which decay algebraically as

$$\chi_{\Delta} \sim N^{-\beta_2}, \quad (13)$$

$$\chi_r \sim N^{-1/\nu_2}. \quad (14)$$

We anticipate a finite-size scaling form of fluctuation distribution functions as

$$P_{\Delta}(\delta\Delta, N) = N^{\beta_2} f_1(\delta\Delta N^{\beta_2}), \quad (15)$$

$$P_r(\delta r_c, N) = N^{1/\nu_2} f_2(\delta r_c N^{1/\nu_2}). \quad (16)$$

The universality class of continuous clique percolation is characterized by the critical exponents  $\beta_1$ ,  $\beta_2$ ,  $\nu_1$  and  $\nu_2$ . Different clique percolation with the same critical exponents belong to the same universality class.

In Fig.2(a), the average  $\bar{\Delta}$  is plotted with respect to network size  $N$  for  $(k, l) = (3, 1), (3, 2), (4, 1), (4, 2)$ , and  $(4, 3)$ . The log-log plot of  $\bar{\Delta}$  versus  $N$  show that  $\bar{\Delta} \propto N^{-\beta_1}$ . From the slope of fitting line, we can get the exponent  $\beta_1$ . It has been obtained that  $\beta_1 = 0.331(5)$  for  $(2, 1)$ ,  $\beta_1 = 0.32(2)$  for  $(3, 1)$ ,  $\beta_1 = 0.33(1)$  for  $(4, 1)$ , and  $\beta_1 = 0.33(3)$  for  $(5, 1)$ . Within error bars,  $\beta_1$  for  $l = 1$  and different  $k$  are equal. At  $l = 2$ , we get  $\beta_1 = 0.05(2)$  for  $(3, 2)$  and  $\beta_1 = 0.04(2)$  for  $(4, 2)$ . The exponents  $\beta_1$  of  $l = 2$  are different from that of  $l = 1$ . At  $l = 3$ ,  $\beta_1 = 0.01(1)$  for  $(4, 3)$  and is different from that of  $l = 1$  and  $l = 2$ . So the universality class of  $(k, l)$  clique percolation is characterized by  $l$ . In Fig.2(b), the log-log plot of  $\chi_{\Delta}$  versus  $N$  is shown for different  $(k, l)$ . The slope of curve gives the exponent  $\beta_2$ . It is found that  $\beta_2$  is equal to the corresponding  $\beta_1$  within error bars.

The finite-size effects of  $\bar{r}_c(N) - r_c(\infty)$  are shown in Fig. 3(a) for  $l < k-1$  and Fig. 3(b) for  $l = k-1$ . From

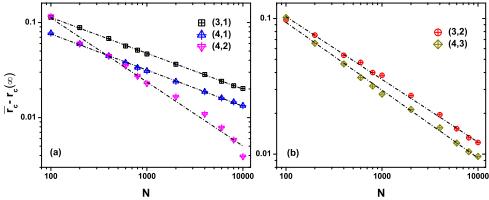


FIG. 3: (color online) Log-log plot of  $\bar{r}_c - r_c(\infty)$  versus  $N$ . The results of  $l < k - 1$  is shown in (a) and  $l = k - 1$  in (b). The slopes of fitting lines give  $1/\nu_1$ , which is defined in Eq. 9 and summarized in Table I.

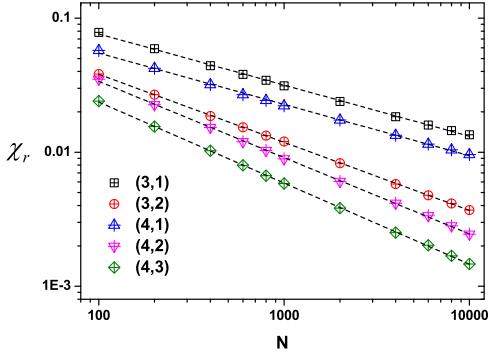


FIG. 4: (color online) Log-log plot of  $\chi_r$  versus  $N$ . The slopes of fitting lines give  $1/\nu_2$ , which is defined in Eq. 12 and summarized in Table I.

the Monte Carlo data, the values of  $r_c(\infty)$  and  $\nu_1$  are determined simultaneously. For  $l = k - 1$ , we get  $1/\nu_1 = 0.46(2)$  for  $(3, 2)$  and  $1/\nu_1 = 0.49(4)$  for  $(4, 3)$ . For  $l < k - 1$ , we obtain  $1/\nu_1 = 0.37(2)$  for  $(3, 1)$ ,  $1/\nu_1 = 0.38(2)$  for  $(4, 1)$ , and  $1/\nu_1 = 0.67(20)$  for  $(4, 2)$ . The  $\bar{r}_c(N)$  of  $(4, 2)$  clique percolation shows deviation from a simple power law of  $N$ . This deviation could be resulted by additional finite-size terms.  $1/\nu_1$  of different  $k$  and the same  $l$  are equal within error bars. We summarize  $r_c(\infty)$  and  $1/\nu_1$  of different  $(k, l)$  clique percolation in Table I.

The root-mean-square of  $\delta r_c$  is shown in Fig. 4. Our results of simulations confirm the power-law behavior of  $\chi_r$ . For  $k = 3$ , we get  $1/\nu_2 = 0.33(5)$  for  $l = 1$  and  $1/\nu_2 = 0.50(2)$  for  $l = 2$ . At  $k = 4$ , we obtain  $1/\nu_2 = 0.35(3)$  for  $l = 1$ ,  $1/\nu_2 = 0.52(4)$  for  $l = 2$ , and  $1/\nu_2 = 0.60(3)$  for  $l = 3$ . Other results of  $1/\nu_2$  are given in Table I.  $1/\nu_2$  of different  $k$  and the same  $l$  are equal within error bars. This confirms further that the universality of  $(k, l)$  clique percolation is characterized alone by  $l$ . The exponent  $\nu_2$  agrees with  $\nu_1$  at  $l = 1, 2$ , but differs from  $\nu_1$  at  $l = 3$ .

For a network with size  $N$ , we simulate its evolution for many times. At each simulation, we can get the largest reduced size gap  $\Delta$  and the corresponding transition point  $r_c$ . From the results of many simulations, the fluctuation distribution functions  $P_\Delta(\delta\Delta, N)$  and  $P_r(\delta r_c, N)$  can be obtained. Five different network sizes  $N = 100, 600, 1000, 6000, 10000$  are chosen. Each fluctuation distribution function has five different curves. Using the finite-

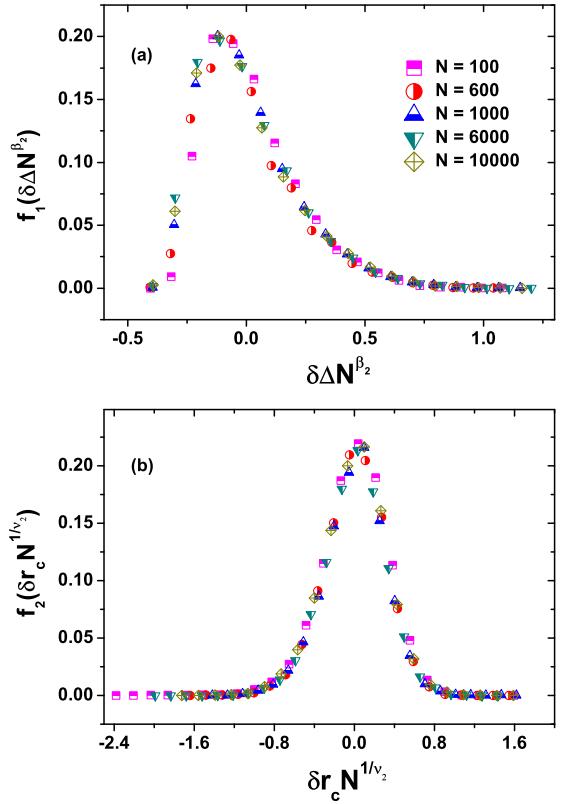


FIG. 5: (color online) Finite-size scaling functions of fluctuation distribution functions (a)  $P_\Delta(\delta\Delta, N)$  and (b)  $P_r(\delta r_c, N)$  for  $(4, 1)$  clique percolation.

size scaling forms of Eq. 15 and 16, five curves of each distribution function collapse into one curve of finite-size scaling function. The scaling variables are defined with the exponents  $\beta_2$  and  $\nu_2$  obtained from  $\chi_\Delta$  and  $\chi_r$ . In Fig. 5, the finite-size scaling functions  $f_1(\delta\Delta N^{\beta_2})$  and  $f_2(\delta r_c N^{1/\nu_2})$  are shown for  $(4, 1)$  clique percolation.  $f_1$  and  $f_2$  of other  $(k, l)$  clique percolation have similar behavior.

In the evolution process of the ER model beginning from  $N$  isolated nodes to being fully connected with  $N(N - 1)/2$  vertices, there are a series of  $(k, l)$  clique percolation. The transition point of a  $(k, l)$  clique percolation is  $T_c = \frac{1}{2}a(k, l)N^{2-\frac{2}{k+l-1}}$ , which is of the same order of  $N$  for different clique percolation with equal  $k + 1$ . The sequence of clique percolation in the ER model is as following: a)  $(2, 1)$  clique percolation at  $T_c = 0.5N$ , b)  $(3, 1)$  clique percolation at  $T_c = 0.5N^{4/3}$ , c)  $(3, 2)$  clique percolation at  $T_c = 2^{-3/2}N^{3/2} \simeq 0.354N^{3/2}$ , d)  $(4, 1)$  clique percolation at  $T_c = 2^{-5/6}N^{3/2} \simeq 0.561N^{3/2}$ , e)  $(4, 2)$  clique percolation at  $T_c = 2^{-1}2.5^{-1/5}N^{8/5} \simeq 0.416N^{8/5}$ , f)  $(5, 1)$  clique percolation at  $T_c = 2^{-1}6^{1/10}N^{8/5} \simeq 0.598N^{8/5}$ , g)  $(4, 3)$  clique percolation at  $T_c = 2^{-1}3^{-1/3}N^{5/3} \simeq 0.347N^{5/3}$ , and so on.

For comparison with the investigations above, we investigate directly the size of the largest  $(k, l)$  clique community in a network to study general clique percolation. For a network

TABLE I: Summary of reduced transition points  $r_c = T_c/N^{2-\frac{2}{k+l-1}}$  and critical exponents of  $(k, l)$  clique percolation.  $r_c(\infty)$  is obtained from Eq. 9 and the analytic result  $r_c^T(\infty)$  is given in Eq. 10. The results of  $(2, 1)$  clique percolation are taken from[14].

$(k,l)$	$r_c(\infty)$	$r_c^T(\infty)$	$\beta_1$	$\beta_2$	$1/\nu_1$	$1/\nu_2$
$(2,1)$	0.5006(7)	$2^{-1}$	0.331(5)	0.334(2)	0.331(7)	0.334(3)
$(3,1)$	0.50(1)	$2^{-1}$	0.32(2)	0.36(4)	0.37(2)	0.33(5)
$(3,2)$	0.34(2)	$2^{-3/2}$	0.05(2)	0.03(2)	0.46(2)	0.50(2)
$(4,1)$	0.56(1)	$2^{-5/6}$	0.33(1)	0.35(2)	0.38(2)	0.35(3)
$(4,2)$	0.40(1)	$2^{-1} \cdot 2.5^{-1/5}$	0.04(2)	0.03(2)	0.67(20)	0.52(4)
$(4,3)$	0.34(2)	$2^{-1} \cdot 3^{-1/3}$	0.01(1)	0.01(1)	0.49(4)	0.60(3)
$(5,1)$	0.60(1)	$2^{-1} \cdot 6^{1/10}$	0.33(3)	0.38(4)	0.30(5)	0.37(6)

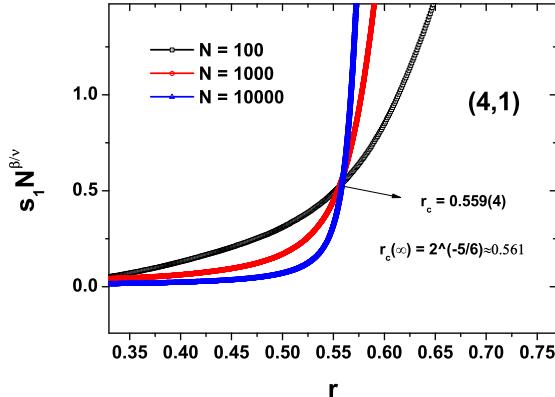


FIG. 6: (color online) Scaling size of the largest  $(4, 1)$  clique community  $s_1 N^{\beta/\nu}$  versus the reduced edge number  $r = T/N^{3/2}$ . We take  $\beta/\nu = 0.33$ . There is a fixed point at  $r_c = 0.559(4)$ .

with  $N$  vertices and  $T$  edges, the reduced size of the largest  $(k, l)$  clique community is denoted as  $s_1(T, N)$ . Near the transition point  $T_c$  of a  $(k, l)$  clique percolation, we anticipate a finite-size scaling form

$$s_1(T, N) = N^{-\beta/\nu} \tilde{s}_1(tN^{1/\nu}), \quad (17)$$

where  $t = (T - T_c)/T_c$  and  $\nu$  is the critical exponent of correlation length. In the limit  $N \rightarrow \infty$ ,  $s_1(T, \infty) = 0$  at  $T < T_c$  and  $s_1(T, \infty) = A_1 t^\beta$  at  $T > T_c$ . Our previous investigations [12, 13] has confirmed the finite-size scaling form Eq. 17 for  $(2, 1)$  clique percolation.

At the transition point  $T_c$ ,  $s_1(T, N)N^{\beta/\nu}|_{T=T_c} = \tilde{s}_1(0)$  and is independent of network size  $N$ . Using this property, we can determine the transition point  $T_c$  of  $(k, l)$  clique percolation from the fixed point of  $s_1(T, N)N^{\beta/\nu}$ .

In Fig. 6, we plot  $s_1(T, N)N^{\beta/\nu}$  of  $(4, 1)$  clique community as a function of  $r$  for different  $N$ .  $\beta/\nu = 0.33$  has been taken. We find a fixed point at  $r_c = 0.559(4)$ , which agrees with  $r_c(\infty) = 0.56(1)$  obtained from network evolution. We expect that  $\beta/\nu$  is equal to  $\beta_2$  [14].

In summary, we introduce a  $(k, l)$  clique community consisting of adjacent  $k$ -cliques sharing at least  $l$  vertices with  $k - 1 \geq l \geq 1$ . There is a  $(k, l)$  clique percolation with

the emergence of a giant  $(k, l)$  clique community. We study the  $(k, l)$  clique percolation by investigating the largest reduced size gap  $\Delta$  of the largest clique community during network evolution and the corresponding evolution step  $T_c$ . If the average size gap  $\bar{\Delta} \sim N^{-\beta_1}$ , there is a continuous clique percolation for  $0 < \beta_1 < 1$  and a discontinuous clique percolation when  $\beta_1 = 0$ . The reduced transition point  $\bar{r}_c$  obtained from  $T_c$  has the finite-size effect  $\bar{r}_c - r_c(\infty) \sim N^{-1/\nu_1}$ . The values of  $r_c(\infty)$  obtained from Monte Carlo data agree with the analytic result of Eq. 5, which is derived using generating function method. The sequence of  $(k, l)$  clique percolation in the ER model is as following:  $(2, 1), (3, 1), (3, 2), (4, 1), (4, 2), (5, 1), (4, 3)$ , and so on. The root-mean-squares of fluctuations  $\delta\Delta$  and  $\delta r_c$  are found to decay algebraically as  $\chi_\Delta \sim N^{-\beta_2}$  and  $\chi_r \sim N^{-1/\nu_2}$ . The fluctuation distribution functions follow the finite-size scaling forms  $P_\Delta(\delta\Delta, N) = N^{\beta_2} f_1(\delta\Delta N^{\beta_2})$  and  $P_r(\delta r_c, N) = N^{1/\nu_2} f_2(\delta r_c N^{1/\nu_2})$ , respectively.

The universality of  $(k, l)$  clique percolation is characterized by the critical exponents  $\beta_1, \beta_2, \nu_1$  and  $\nu_2$ . It has been found that the critical exponents of  $(k, l)$  clique percolation are independent of  $k$ , but dependent on  $l$ . The universality class of  $(k, l)$  clique percolation is characterized alone by  $l$ .

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\* Electronic address: chenxs@itp.ac.cn

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